Response to Reviewer 1's comments on "Microbial dynamics in a High-Arctic glacier forefield: a combined field, laboratory, and modelling approach."

We are thankful for the helpful comments of the reviewer in the third round of revisions. We have addressed all final concerns that were raised.

The authors have made a sincere and detailed effort to improve the manuscript by addressing comments made in the earlier reviews. I am still wary about the use of measurements at 25*C with soils conditioned to 5*C, but can accept its necessity for the present study. The open nature of the review process offers readers with the inclination an opportunity to view the arguments for and against the use of 25*C temperatures.

Sequencing 16S rRNA genes to parameterize the model.
I still have problems here. It is assumed that shifts in the absolute quantities of different functional/taxonomic groups are meaningfully represented by DNA extracted from varying quantities of samples, amplified and sequenced to an unstated but incomplete depth, used to generate relative abundance profiles of sequence libraries which may differ in read number, and in any case are based upon sequences from a locus which has variable numbers of copies per bacterial genome, a source of variation which is known to be related to ecological traits important in glacier forefield successional processes and itself recapitulated within the data. Specifically:

A. L179: Thanks for updating the details here, but I have concerns over "5-10 grams of sample" being used for DNA extraction.
1. If this is the case, the statement DNA was extracted by "MoBio PowerSoil® DNA Isolation Kit and by following the instruction manual cannot be the case. This kit processes samples in 2 mL tubes for bead beating. The standard recommendation from the manufacturer is for 250 mg samples (and I have had no problems with getting enough DNA from forefield soils for amplicon sequencing from 250 mg using the standard protocol.) The key point is that 5-10 grams would not fit, and any more than 250-500 mg clogs the bead beating tube to the point that extraction efficiency is reduced, making using more sample counterproductive, so I think there's a mistake here. Did the authors actually use a PowerMax kit which can process up to 10 g of sample in a very different format, or modify the powersoil protocol to pool extracts from multiple 250 mg extracts? This may seem a trivial detail, but establishing how you extracted the DNA is important to establish whether the sequencing data have any relation to the soil community.

We would like to thank the referee for this important point. The "MoBio PowerSoil® DNA Isolation Kit" as referenced in the manuscript was indeed a mistake. We have corrected this to "PowerMax ® Soil DNA Isolation Kit" in the revised manuscript (line 178).

2. So, some of your samples were 5 grams and some were 10 grams? How did you account for up to two fold variation in sample mass, with the implications for the total biomass in the tube, and furthermore the extractable biomass as a result of variation in extraction efficiency from the consequently variable ratios and mixing in volumes of sample, beads, buffer and tube during bead beating?

The referee correctly points out that by using 5-10g for the DNA extraction there could have been up to two fold variation in sample mass per tube, resulting in variation in the total biomass in the tube. It is worth noting that variation in biomass in each extraction tube also occurs regardless of sample volume, due to natural variability in cell biomass abundance along the chronosequence (see Table 2).

In an effort to normalize between samples with varying amounts of biomass following extraction, DNA was diluted to equimolar concentrations, so that all samples contained 7.0 ng of DNA prior to amplification. Then, following amplification, all amplicons were normalized to the same concentration (9.0 ng) before sequencing. This information has been added to the revised manuscript (lines 180-190).

B. L193-200: To establish the depth of sequencing and how many reads were there per barcode, and was the depth of sequencing? In other words, how can the reader be assured that your sequencing effort meaningfully describes the community, and that your relative abundance profiles are from roughly equivalent (or rarefied) numbers of reads. This matters because you are using proportional data from sequencing libraries to make estimations of absolute contributions to the soil community your model describes.

The final OTU table contained between 2329-43269 reads for each sample. The varying sequence coverage among samples was adjusted by normalizing the OTU matrix to the same total abundance of sequences for each sample (2329) using the command normalize_table.py in QIIME. Rarefaction curves (below) show incomplete sample depth. However, despite shortcomings of the depth of sequencing, the taxonomic information provides a useful snapshot with which to support and complement the modelling exercise and does not affect the major conclusions of the study.

This information has been included and discussed in the revised manuscript (lines 200-207, 619-623 and 674-675) and the rarefaction plot has been included in the Supplementary Material (Fig. S6) for transparency.

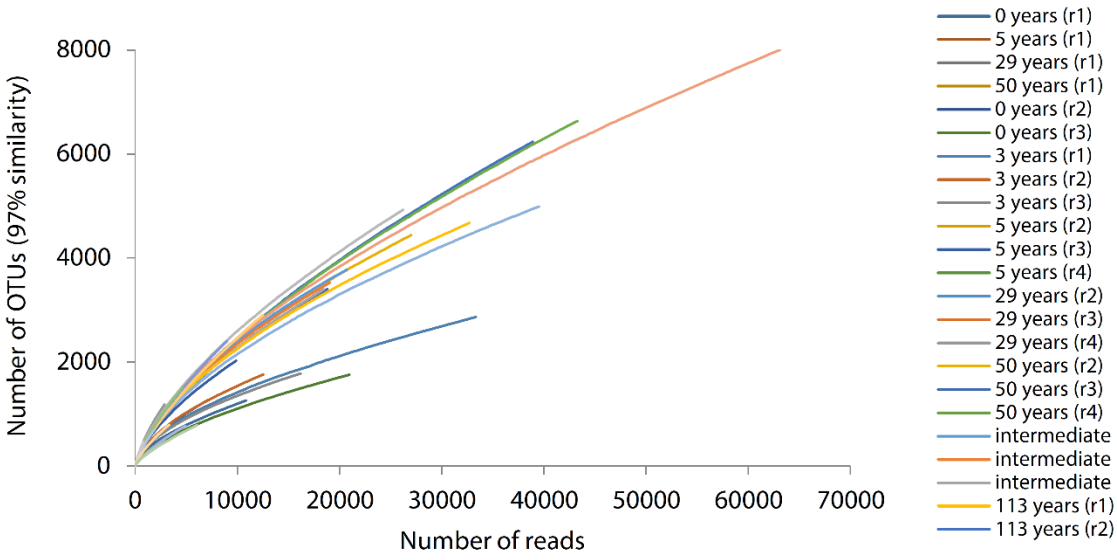

Fig. S6. Rarefaction curves for sequencing.

C. From the taxonomic profiles I see important changes in the community's composition and configuration which mirror other work on transitions between

oligotrophic and copiotrophic taxa in a variety of glacier chronosequences. That transition is associated with changes in ribosomal RNA gene copy number per genome. There are bioinformatic means of normalising this variation in rRNA copy number, for example CopyRighter (DOI: 10.1186/2049-2618-2-11).

We thank the reviewer for bringing up this concern. Following this, we have implemented CopyRighter on some taxa, and results were not different. However, implementing on all taxa and re-classifying for the functional groups as defined by the model is not possible at this stage. Thus, as the reviewer has suggested below, we have added a caveat to the discussion to state that absolute numbers should be treated with caution due to the differential extractability of DNA from different organisms and variation in rDNA copy number (lines 619-622 and 668-670).

I would hope the authors can state succinctly the measures taken to address problems A and B, and caveat their discussion to address concern C. This would serve to improve this paper, and perhaps consideration of these issues will help the other paper mentioned in their response which will probably rely on this 16S data a central pillar. As a general thought, for future endeavours and considering the broad taxonomic resolution desired by the authors to inform their model, I feel there are more appropriate tools for this job. If DNA based analyses were needed, qPCR or FISH would have provided data more closely linked to microbial abundance. Otherwise, PLFA would have provided taxonomically-informed estimates of contributions to biomass.

We hope that the above has satisfied the reviewer's concerns regarding the genomic data component of the study.

Minor comments.

L180: At this point it is not isolated rDNA; it will be community genomic DNA, from which you will specifically amplify bacterial 16S ribosomal RNA genes (rDNA itself is a common misnomer too, by the way: the gene product is rRNA, hence rRNA gene)

Corrected (line 181)

L183: dNTPs

Corrected (line 184)

Accession numbers: Hopefully NCBI will have returned the numbers soon.

Sequence data is available on NCBI under project ID PRJNA341831 (lines 205-207), and individual accession numbers are provided in the Supplementary Information.

Accession nr.	Sample Name	Soil Age
SAMN05729451	S132101a	0
SAMN05729452	S132102a	0
SAMN05729453	S131101a	0
SAMN05729454	S132301a	29
SAMN05729455	S132303a	29
SAMN05729456	S132302a	29
SAMN05729457	S131301a	29
SAMN05729458	S133301a	29

SAMN05729459	S132503a	50
SAMN05729460	S132501a	50
SAMN05729461	S132502a	50
SAMN05729462	S131501a	50
SAMN05729463	S133501a	50
SAMN05729464	S132603a	113
SAMN05729465	S132602a	113
SAMN05729466	S132601a	113

Response: " Assuming that predation rates and dormancy are the same on all functional groups (as is defined in SHIMMER)," I advise caution in making that assumption as these traits are varied across the breadth of soil microbiota.

We acknowledge that the assumption of equal predation/dormancy rates between functional groups is a simplification, but it is a necessary assumption for the purpose of model-16S comparison using this method. We are thankful to the reviewer for suggesting the use of other techniques (qPCR, FISH, PLFA) for future endeavours.

Microbial dynamics in a High-Arctic glacier forefield: a combined field, laboratory, and modelling approach.

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Abstract: Modelling the development of soils in glacier forefields is necessary in order to assess how microbial and geochemical processes interact and shape soil development in response to glacier retreat. Furthermore, such models can help us predict microbial growth and the fate of Arctic soils in an increasingly ice-free future. Here, for the first time, we combined field sampling with laboratory analyses and numerical modelling to investigate microbial community dynamics in oligotrophic proglacial soils in Svalbard. We measured low bacterial growth rates and growth efficiencies (relative to estimates from Alpine glacier forefields), and high sensitivity to soil temperature (relative to temperate soils). We used these laboratory measurements to inform parameter values in a new numerical model and significantly refined predictions of microbial and biogeochemical dynamics of soil development over a period of roughly 120 years. The model predicted the observed accumulation of autotrophic and heterotrophic biomass. Genomic data indicated that initial microbial communities were dominated by bacteria derived from the glacial environment, whereas older soils hosted a mixed community of autotrophic and heterotrophic bacteria. This finding was simulated by the numerical model, which showed that active microbial communities play key roles in fixing and recycling carbon and nutrients. We also demonstrated the role of allochthonous carbon and microbial necromass in sustaining a pool of organic material, despite high heterotrophic activity in older soils. This combined field, laboratory and modelling approach demonstrates the value of integrated model-data studies to understand and quantify the functioning of the microbial community in an emerging High-Arctic soil ecosystem.

Key words
Glacier forefield
Microbial dynamics
Soil development
Numerical modelling
Integrated field-laboratory-modelling
SHIMMER

1. Introduction

Polar regions are particularly sensitive to anthropogenic climate change (Lee, 2014) and have experienced accelerated warming in recent decades (Johannessen et al., 2004; Serreze et al., 2000; Moritz et al., 2002). The response of terrestrial Polar ecosystems to this warming is complex, and research to understand the response of terrestrial ecosystems in high latitudes to environmental change is of increasing importance. A visible consequence of Arctic warming is the large-scale retreat of glacier and ice cover (ACIA, 2005; Paul et al., 2011; Staines et al., 2014; Dyurgerov and Meier, 2000). From underneath the ice, a new terrestrial biosphere emerges, playing host to an ecosystem which may exert an important influence on biogeochemical cycles, and more specifically atmospheric CO_2 concentrations and associated climate feedbacks (Dessert et al., 2003; Anderson et al., 2000; Smittenberg et al., 2012; Berner et al., 1983).

Numerous studies have attempted to characterize the physical and biological development of recently exposed soils using a chronosequence approach, whereby a transect perpendicular to the retreating ice snout represents a time sequence with older soils at increasing distance from the ice snout (Schulz et al., 2013). We have recently shown that microbial biomass and macronutrients (such as carbon, phosphorus and nitrogen) can accumulate in soils over timescales of decades to centuries (Bradley et al., 2014). In such pristine glacial forefield soils the activity of microbial communities is thought to be responsible for this initial accumulation of carbon and nutrients. Such an accumulation facilitates colonization by higher order plants, leading to the accumulation of substantial amounts of organic carbon (Insam and Haselwandter, 1989). However, organic carbon may also be derived from allochthonous sources such as material deposited on the soil surface (from wind, hydrology, precipitation and ornithogenic sources) and ancient organic pools derived from under the glacier (Schulz et al., 2013). Nevertheless, the relative significance of allochthonous and autochthonous sources of carbon to forefield soils, as well as their effect on ecosystem behaviour are so far still poorly understood (Bradley et al., 2014). Moreover, cycling of bioavailable nitrogen (which is derived from active nitrogen-fixing organisms, allochthonous deposition, and degradation of organic substrates) and phosphorus (liberated from the weathering of minerals and decomposition of organic substrates) are similarly poorly quantified.

Several studies have observed shifts in the microbial community inhabiting pro-glacial soils of various ages (Zumsteg et al., 2012; Zumsteg et al., 2011). This was expressed in increasing rates of autotrophic and bacterial production with soil age (Schmidt et al., 2008; Zumsteg et al., 2013; Esperschutz et al., 2011; Frey et al., 2013) and the overall decline in quality of organic substrates in older soils (Goransson et al., 2011; Insam and Haselwandter, 1989). However, current evidence is limited to mostly descriptive approaches, which may be challenging to interpret due to inherent difficulties in disentangling interacting microbial and geochemical processes across various temporal and spatial scales. Furthermore, the inherent heterogeneity of glacial forefield soils makes the development of a single conceptual model that fits all challenging. Accordingly, pro-glacial biogeochemical processes that dominate such systems

remain poorly quantified and highly under-explored. This current lack of understanding limits our ability to predict the future evolution of these emerging landscapes and the potential consequences on global climate. Numerical models present an opportunity to expand our knowledge of glacier forefield ecosystems by analytically testing the hypotheses that arise from observations, extrapolating, interpolating and budgeting processes, rates and other features to explore beyond the possibility of empirical observation (Bradley et al., 2016). With such a model we can then also explore the sensitivity and resilience of these ecosystems to environmental change.

Here, we have combined field observations, with laboratory incubations and elemental measurements as well as genomic analyses and used these in a numerical model to investigate the development of soils in a glacial forefield. The present study forms an important part of the integrated and iterative model-data approach outlined in the model description and testing (Bradley et al, 2015) whereby initial model development was informed by decades of empirical research, new data and laboratory experiments (presented here) are used to refine and inform model simulations, and so forth. With this data we refined some model parameters in the recently developed ***S***oil biogeoc***H***em***I***cal ***M***odel for ***M***icrobial ***E***cosystem ***R***esponse (SHIMMER 1.0; Bradley et al. (2015)) model and applied this to the emerging forefield of the Midtre Lovénbreen glacier in Svalbard. Pioneer soils in the High-Arctic and Antarctica, such as the Midtre Lovénbreen forefield, are ideal sites to test this field-laboratory-model approach due to the lack of vegetation during initial stages of soil development, as the presence of vegetation would obscure the microbial community dynamics and considerably alter the physical properties of the soil (Brown and Jumpponen, 2014; Ensign et al., 2006; King et al., 2008; Kastovska et al., 2005; Schutte et al., 2009; Duc et al., 2009). The model development was informed by decades of empirical research on glacier forefield soils, and has already been tested and validated using published datasets from the Damma Glacier in Switzerland and the Athabasca Glacier in Canada. A thorough sensitivity analysis highlighted the most important parameters to constrain in order to make further predictions more robust. All our model parameter values are specific to individual, local model conditions and inherently contain necessary model simplifications, abstractions and assumptions. Nevertheless, our earlier sensitivity analyses revealed the following highly sensitive key parameters as the most important to constrain through measurements: the maximum heterotrophic growth rate (I_{maxH}), the bacterial growth efficiency (BGE, parameter Y_H) and the temperature response (Q_{10}).

Therefore, in this current study, we combined detailed field measurements with specifically designed laboratory experiments and quantified values for these three parameters with a specific set of soils from for the Midtre Lovénbreen forefield. The laboratory experiments and measurements were conducted with the objective to better constrain these sensitive parameters. We then ran model simulations in order to explore the ranges of model output and refine model predictions compared to the previous range identified in Bradley et al (2015). Next, we examined model output to explore the microbial and biogeochemical dynamics of recently exposed soils in the Midtre Lovénbreen catchment and evaluate two main hypotheses. First, we tested the hypothesis that microbial biomass in recently exposed soils accumulates due to *in situ* bacterial growth and activity. It is commonly observed in glacier forefields

that microbial biomass accumulates with increasing soil age following exposure (Bernasconi et al., 2011; Schulz et al., 2013; Bradley et al., 2014). This study provides a new quantitative and process-focused approach to examine *in situ* growth in pioneer ecosystems, and assess the role of different functional groups in biomass accumulation. Second, we tested the hypothesis that carbon fluxes in very recently exposed soils are low, and are dominated by (abiotic) deposition of allochthonous substrate, whereas carbon fluxes are high in older soils due to increased microbial (biotic) activity (such as microbial growth, respiration and cell death). Increased soil carbon fluxes with soil age have been linked to microbial activity from the forefield of the Damma Glacier, Switzerland (Smittenberg et al., 2012; Guelland et al., 2013b). With this combined model, field and lab study, we were able to estimate carbon fluxes between ecosystem components with daily resolution, and provide new insight into the interplay of processes that contribute to net ecosystem production and soil organic carbon stocks in a High-Arctic system.

2. Methods

2.1. Study site and sampling

Midtre Lovénbreen is an Arctic polythermal valley glacier on the south side of Kongsfjorden, Western Svalbard (latitude 78°55'N, longitude 12°10'E) (Fig. 1). The Midtre Lovénbreen catchment is roughly 5 km East of Ny-Ålesund, where several long-term monitoring programs have provided a wealth of contextual information. Midtre Lovénbreen has experienced negative mass balance throughout much of the 20th century. Since the end of the Little Ice Age (maximum in Svalbard in the 1900s) the de-glaciated surface area of the Midtre Lovénbreen catchment has increased considerably in response to warming mean annual temperatures. This continues to the present day. Between 1966 and 1990 ~ 2.3 km^2 of land was exposed (Fleming et al., 1997; Moreau et al., 2008). We used a chronosequence approach to determine ages for soils based on satellite imagery (Landsat TM 7) and previously determined soil ages by aerial photography and carbon-14 dating techniques in Hodkinson et al. (2003). Soil samples were collected along a transect perpendicular to the glacier snout, representing soil ages of 0, 3, 5, 29, 50, and 113 years (Fig. 1) during the field season (18 July to 29 August 2013). At each of the 6 sites along the chronosequence, 10 meter traverses roughly parallel to the glacier snout were established and at each site 3 soil plots were sampled (using ethanol sterilized sampling equipment). After removing the > 2 cm rock pieces at each site, about 100 grams of soil was collected from the top 15 cm and immediately placed into sterile high-density polyethylene bags (Whirl-Pak (Lactun, Australia)) and into a cool box partially filled with cool packs and dry ice. Samples were immediately frozen and stored at -20°C on return to the UK Arctic Research station in Ny-Ålesund (no longer than 5 hours after collection). Samples were transported frozen on dry ice to the laboratories in the Universities of Bristol and Leeds (UK).

2.2. Laboratory analyses

For bacterial abundance, samples were thawed and aliquots (100 mg) were immediately transferred into sterile 1.5 mL micro-centrifuge (Eppendorf) tubes, where they were diluted with 900 µL of Milli-Q water (0.2 µm filtered) and immediately fixed in 100 µL glutaraldehyde (0.2 µm filtered, 2.5% final

concentration). Samples were then vortexed for 10 seconds and sonicated for 1 minute at 30°C to facilitate cell detachment from soil particles. Then 10 µL fluorochrome DAPI (4', 6-diamidino-2 phenylindole) was added to half of the samples, tubes were vortexed briefly (3 seconds) and incubated in the dark for 10 minutes, to be counted under UV light. The other half of each sample remained untreated, for counting under auto-fluorescent light for photosynthetic pigmentation. Samples were vortexed for 10 seconds and let stand for a further 30 seconds to ensure a well-mixed solution, prior to filtering 100 µL of the mixed liquid sample onto black Millipore Isopore membrane filters (0.2 µm pore size, 25 mm diameter), rinsed with a further 250 µL of Milli-Q water (0.2 µm filtered). Bacterial cells were then counted using an Olympus BX41 microscope at 1000 times magnification. The filtering apparatus was washed out with Milli-Q water between each filtration, and negative control samples, prepared using Milli-Q water, were included into each series. A negative control was a sample with no visible stained or auto-fluorescing cells. Thirty random grids (each 10^4 µm^2) were counted per sample. Cell morphologies were measured and cell volume was estimated and converted to carbon content according to Bratbak and Dundas (1984) (see Supplementary Information). Separate aliquots of soil from each site were weighed after thawing and then dried at 105°C to obtain an estimate of soil moisture content.

Environmental DNA was isolated from at least 3 replicates for each soil age using MoBio PowerMax ® Soil DNA Isolation Kit~~PowerSoil® DNA Isolation Kit~~ and by following the instruction manual. 5 to 10 g of soil was used per sample to isolate DNA. Following extraction, DNA was diluted to equimolar concentrations, so that all samples contained 7.0 ng of DNA, prior to amplification. The isolated ~~16S~~ ~~r~~genomic DNA was amplified with bacterial primers 515f (5'-GTGYCAGCMGCCGCGGTAA-3') and 926r (5'-CCGYCAATTYMTTTRAGTTT-3') (Caporaso et al., 2012), creating a single amplicon of ~400 bp. The reaction was carried out in 50 µL volumes containing 0.3 mg mL^{-1} Bovine Serum Albumin, 250 µM d~~T~~NTPs, 0.5 µM of each primer, 0.02 U Phusion High-Fidelity DNA Polymerase (Finnzymes OY, Espoo, Finland) and 5x Phusion HF Buffer containing 1.5 mM $MgCl_2$. The following PCR conditions were used: initial denaturation at 95°C for 5 minutes, followed by 25 cycles consisting of denaturation (95°C for 40 seconds), annealing (55°C for 2 minutes) and extension (72°C for 1 minute) and a final extension step at 72°C for 7 minutes. ~~S~~ Following amplification, all amplicons were normalized to the same concentration (9.0 ng) before sequencing. Samples were sequenced using the Ion Torrent platform (using Ion 318v2 chip) at Bristol Genomics facility at the University of Bristol. Samples were barcoded in the PCR stage and demultiplexed in QIIME using split_libraries.py code (Caporaso et al., 2010). A non-barcoded library was prepared from the amplicon pool using Life technologies Short Amplicon Prep Ion Plus Fragment Library Kit. The template and sequencing kits used were: Ion PGM Template OT2 400 Kit and Ion PGM Sequencing 400 kit. The sequencing yielded 4.38 million reads. The 16S sequences were further processed using MOTHUR (v. 1.35) and QIIME pipelines (Schloss et al., 2009; Caporaso et al., 2010). Initially, low quality or too long and too short sequences were removed in MOTHUR. Chimeric sequences were identified and removed using UCHIME (Edgar et al., 2011). QIIME was used to cluster reads into operational taxonomical units (OTUs) using the pick_closed_reference_otus.py command. The final OTU table contained between 2329-43269 reads

for each sample. The varying sequence coverage among samples was adjusted by normalizing the OTU matrix to the same total abundance of sequences for each sample (2329) using the command normalize_table.py in QIIME. Sequences were clustered into OTUs based on at least 97% sequence similarity, and assigned taxonomical identification against Greengenes bacterial database (McDonald et al., 2012). The result was a biom format file containing the taxonomic information for each OTU as well as OTU frequency per sample. Sequence data is available on NCBI under project ID PRJNA341831, and individual accession numbers are provided in the Supplementary Information.

The carbon contents in the year 0 soils were analysed with a Carlo-Erba elemental analyser (NC2500) at the German Research Center for Geosciences, Potsdam, Germany. The soils were oven dried at 40°C for 48 hours, sieved to <7 mm and crushed using a TEMA disk mill to achieve size fractions of < 20 µm. Total organic carbon (TOC) was analysed after reacting the powders with a 10% HCl solution for 12 hours to remove inorganic carbonates.

2.3. Determination of maximum growth rates

The microbial activity was determined in 113 year old soil samples after they were thawed (in the dark at 5°C to mimic typical field temperature) for 168 hours. This age was chosen because these soil samples were assumed to be the ones with the highest microbial biomass and activity and thus the most practical for all laboratory measurements. In order to mitigate the effect of variability derived from differences in soil properties between soil ages (that will later be predicted by the model), laboratory experiments were conducted on a single soil age, with replicate incubations to assess the possible variability in rates (and thus parameter values) that can be attributed to experimental procedures and measurement techniques.

Aliquots of the soils were divided into petri dishes (25 g of soil (wet weight) into each petri dish) for subsequent treatments. In order to alleviate nutrient limitations and measure maximum growth rates, four different nutrient conditions were simulated: (1) no addition of nutrients, (2) low (0.03 mg C g^{-1}, 0.008 mg N g^{-1}, 0.02 mg P g^{-1}), (3) medium (0.8 mg C g^{-1}, 0.015 mg N g^{-1}, 0.1 mg P g^{-1}) and (4) high additions (2.4 mg C g^{-1}, 0.024 mg N g^{-1}, 0.3 mg P g^{-1}). The ranges and concentrations were informed by similar experiments in recently exposed proglacial soils at the Damma Glacier, Switzerland (Goransson et al., 2011). Nutrients ($C_6H_{12}O_6$ for C, NH_4NO_3 for N and KH_2PO_4 for P) (Sigma, quality ≥99.0%) were dissolved in 2 mL Milli-Q water (0.2 µm filtered), and mixed into the soils using an ethanol-sterilized spatula. Samples were incubated in the dark for a further 72 hours with the lids on at 25°C, the reference temperature (T_{ref}) at which all rates are defined in SHIMMER prior to adjustment with the temperature dependency expression (using Q_{10}) (Bradley et al., 2015). In order to derive a value for I_{maxH}, we were obligated to perform growth incubations at T_{ref} (25°C) despite this being a more typical soil temperature of Alpine soils rather than High-Arctic soils (see Fig. S3 (c)). However, we are confident that by deriving a Q_{10} value based on incubations of the same soils encapsulating typical (5°C) to high (25°C) soil temperatures, we can numerically derive appropriate actual growth rates from the maximum growth rate (at T_{ref}). We are confident that the major outcomes and conclusions of this study are not

affected by high incubation temperatures since measured growth rates at high temperatures are appropriately scaled using the Q_{10} formulation as measured experimentally. Throughout the whole incubation time, at 24 hour intervals, additional 2 mL aliquots of Milli-Q water (0.2 µm filtered) were added to maintain approximate soil moisture conditions in each sample.

In these samples bacterial production was estimated by the incorporation of ^{3}H-leucine using the microcentrifuge method detailed in Kirchman (2001). After the initial 72 hour incubation period quadruplicate sample aliquots from the petri dish incubations and two trichloroacetic acid (TCA) killed control samples were incubated for 3 hours at T_{ref} (25°C) for every nutrient treatment. Approximately 50 mg of soil was transferred to sterile micro-centrifuge tubes (2.0 mL, Fischer Scientific). Milli-Q (0.2 µm pre-filtered) water and ^{3}H-leucine was added to a final concentration of 100 nM (optimum leucine concentration was pre-determined by a saturation experiment, Fig. S1, Supplementary Information). The incubation was terminated by the addition of TCA to each tube. Tubes were then centrifuged at 15,000 g for 15 minutes, the supernatant was aspirated with a sterile pipette and removed, and 1 ml ice-cold 5% TCA was added to each tube. Tubes were then centrifuged again at 15,000g for 5 minutes, before again aspirating and removing the supernatant. 1mL ice-cold 80% ethanol was added and tubes were centrifuged at 15,000 g for 5 minutes, before the supernatant was aspirated and removed again and tubes were left to air dry for 12 hours. Finally, 1 mL of scintillation cocktail was added, samples were vortexed, and then counted by liquid scintillation (Perkin Elmer Liquid Scintillation Analyzer, Tri-Carb 2810 TR). Radioisotope activity of TCA-killed control samples was always less than 1.1% of the measured activity in live samples. There was a positive correlation between the amount of sediment added to the tubes and background counts representing disintegrations per minute (DPM). Counts were individually normalized by the amount of sediments (corrected for dry weight) used in each sample to discount for background DPM. Leucine incorporation rates were converted into bacterial carbon production following the methodology of Simon and Azam (1989). Bacterial abundance was estimated from each treatment after the 72 hour incubation period by microscopy. Five samples from each petri dish were counted for each nutrient treatment with negative controls yielding no detectable cells. One-way ANOVA (with post-hoc Tukey HSD) statistical tests were used for evaluations of the variability from the multiple treatments.

2.4. Temperature response

Microbial community respiration was determined by measuring CO_2 gas exchange rates in airtight incubation vials. Soil samples from the 113 year old site were defrosted and divided (25 g wet weight) in petri dishes as above, and 2 mL of Milli-Q water (0.2 µm filtered) was added (to maintain consistency of soil moisture with determination of bacterial production above). Samples were incubated at 5°C (T_1) and 25°C (T_2) in the dark for a further 72 hours. 2mL of 0.2 µm pre-filtered Milli-Q water was added to the T_1 sample (3 mL for T_2) at 24, 48 and 72 hours to maintain approximate soil moisture content. Two separate killed control tests (one furnaced at 450°C for 4 hours, and one autoclaved (3 cycles at 121°C)) were incubated at T_1 and T_2. Quintuple live and killed samples (roughly 1 g) were transferred into cleaned 20 mL glass vials (rinsed in 2% Decon, submersed in 10% HCl for 24 hours, rinsed 3 times

with Milli-Q water and furnaced at 450°C for 4 hours). These were sealed (9°C, atmospheric pressure, ambient CO_2 of 405 ppm) with pre-sterilized Bellco butyl stoppers (pre-sterilized by boiling for 4 hours in 1M sodium hydroxide) and crimped shut with aluminium caps. Sealed vials were then incubated at T_1 and T_2 for 24 hours in darkness. After 24 hours, the headspace gas was removed with a gas-tight syringe and immediately analysed on an EGM4 gas analyser (PP Systems, calibrated using gas standards matching the expected range, precision 1.9%, 2*SE). Empty pre-sterilized vials were also incubated and analysed. Following gas analysis, vials were opened and dried to a constant weight at 105°C to estimate moisture content and thus dry soil weight of these aliquots. Headspace CO_2 change (ppm) was converted to microbial respiration using the ideal gas law (n=PV/RT), assuming negligible changes in soil pore water pH (and therefore CO_2 solubility) during the incubation. CO_2 headspace changes resulting from killed controls and blanks were < 70% of the changes resulting from the incubations at T_1, and <7% of the changes observed at T_2. One-way ANOVA (with post-hoc Tukey HSD) statistical tests were used for comparison of multiple treatments. No significant differences in CO_2 headspace change between killed controls at T_1 and T_2 were detected (P=0.95).

2.5. Microbial Model: SHIMMER

SHIMMER (Bradley et al., 2015) mechanistically describes and predicts transformations in carbon, nitrogen and phosphorus through aggregated components of the microbial community as a system of interlinked ordinary differential equations. The model contains pools of microbial biomass, organic matter and both dissolved inorganic and organic nitrogen and phosphorus (Table 1). It categorizes microbes into autotrophs ($A_{1\text{-}3}$) and heterotrophs ($H_{1\text{-}3}$), and further subdivides these based on 3 specific functional traits. Microbes derived from underneath the glacier (referred to as “glacial microbes”) are termed A_1 and H_1. A_1 are chemolithoautotrophic, obtaining energy from the oxidation and reduction of inorganic compounds and carbon from the fixation of carbon dioxide. In contrast, H_1 rely on the breakdown of organic molecules for energy to support growth. A_2 and H_2 represent autotrophic and heterotrophic microbes commonly found in glacier forefield soils with no “special” characteristics, and will be referred to as “soil microbes”. A_3 and H_3 are autotrophs and heterotrophs that are able to fix atmospheric N_2 gas as a source of nitrogen in cases when dissolved inorganic nitrogen (DIN) stocks become limiting. Available organic substrate is assumed to be derived naturally from dead organic matter and allochthonous inputs. Labile compounds are immediately available fresh and highly reactive material, rapidly turned over by the microorganisms (S_1, ON_1, OP_1). Refractory compounds are less bioavailable and represents the bulk of substrate present in the non-living organic component of soil (S_2, ON_2, OP_2). A conceptual diagram showing the components and transfers of SHIMMER is presented in the Supplementary Information (Fig. S2).

Microbial biomass responds dynamically to changing substrate and nutrient availability (expressed as Monod-kinetics), as well as changing environmental conditions (such as temperature and light). A Q_{10} temperature response function (T_f) is affixed to all metabolic processes including growth rates and death rates (Bradley et al., 2015), thus effectively slowing down or speeding up all life processes as temperature changes (Soetaert and Herman, 2009; Yoshitake et al., 2010; Schipper et al., 2014). Light

limitation is expressed as Monod kinetics. The following external forcings drive and regulate the system's dynamics:

- Photosynthetically-active radiation (PAR) (wavelength of approximately 400 to 700 nm) (W m^{-2}).
- Snow depth (m).
- Soil temperature (°C).
- Allochthonous inputs (µg g^{-1} day^{-1}).

The model is 0-D and represents the soil as a homogeneous mix. Thus, light, temperature, nutrients, organic compounds and microbial biomass are assumed to be evenly distributed.

Soil temperature (at 1 cm depth) for the entire of 2013 is provided by Alfred Wegener Institute for Polar and Marine Research (AWI) from the permafrost observatory near Ny-Ålesund, Svalbard. Similarly, PAR for 2013 are measured at the AWI surface radiation station near Ny-Ålesund, Svalbard. Averaged daily snow depth for 2009 to 2013 is provided by the Norwegian Meteorological Institute (eKlima). Allochthonous nutrient fluxes (inputs and leaching) are estimated based on an evaluation of nutrient budgets of the Midtre Lovénbreen catchment (Hodson et al., 2005) in which budgets for nutrient deposition rates and runoff concentrations are measured over two full summer-winter seasons and residual retention rates (excess of inputs) or depletion rates (excess of outputs) are inferred. The bioavailability of allochthonous material is assumed to be the same as initial material and microbial necromass.

Initial conditions were informed by analysis of 0-years-of-exposure soil collected adjacent to the ice snout, and initial values for all state variables are presented in Table 1. Initial microbial biomass was estimated by microscopy as described above. Initial community structure was derived by 16S analysis of year-0 soils. An initial value for carbon substrate (S_1 + S_2) was estimated based on the average TOC content of year-0 soil. Bioavailability of model TOC was assumed to be 30% labile (S_1) and 70% refractory (S_2) (for consistency with Bradley et al. (2015)). Organic nitrogen (ON) and organic phosphorus (OP) were assumed to be stoichiometrically linked by the measured C:N:P ratio from the Damma Glacier forefield (from which the model was initially developed and tested (Bradley et al., 2015)). An initial value for DIN was taken from a previous evaluation of Svalbard tundra nitrogen dynamics, whereby the lowest value is taken to represent the soil of least development, according to traditional understanding of glacier forefields (Alves et al., 2013; Bradley et al., 2014). An initial value for dissolved inorganic phosphorous (DIP) was established stoichiometrically from previous model development and testing.

Model implementation and set-up is described in more detail in the Supplementary Information.

2.6. Model parameters

Maximum heterotrophic growth rate I_{maxH} (day^{-1}) was estimated by scaling the measured rate of bacterial production (µg C g^{-1} day^{-1}) (converted to dry weight) with total heterotrophic biomass (µg C g^{-1}). Nutrient addition alleviates growth limitations as defined in SHIMMER (Bradley et al., 2015); thus bacterial communities can be assumed to be growing at I_{maxH} under experimental conditions.

Y_H represents heterotrophic BGE, and was estimated according to the equation:

$$Y_H = \frac{BP}{BP + BR} \quad (1)$$

Where *BP* is and *BR* are measured bacterial production and measured bacterial respiration (µg C g^{-1} day^{-1}) respectively, at 25°C with no nutrients added.

The temperature response (Q_{10}) value was estimated as:

$$Q_{10} = \left(\frac{R_2}{R_1}\right)^{\left(\frac{10}{T_2 - T_1}\right)} \quad (2)$$

Where R_1 and R_2 represent the measured respiration rate (µg C g^{-1} day^{-1}) at temperatures T_1 and T_2 (5°C and 25°C).

Laboratory-defined parameters (i.e. growth rate, temperature sensitivity and BGE) were assumed to be the same for all microbial groups. A complete list of parameters and values is presented in Table S3 (Supplementary Information).

3. Results

3.1. Laboratory results and model parameters

Bacterial production in untreated soil was estimated at 0.76 µg C g^{-1} day^{-1} (SD=0.12), and across all nutrient treatments ranged from 0.560 to 2.196 µg C g^{-1} day^{-1}. Nutrient addition led to increased measured production (low = 0.69 µg C g^{-1} day^{-1} (SD=0.12), medium = 1.09 µg C g^{-1} day^{-1} (SD=0.53), high = 1.52 µg C g^{-1} day^{-1} (SD=0.63)), however variability between replicates was also high and production rates from each nutrient treatment were not significantly different from untreated soil (P_{low}=0.99, P_{medium}=0.70, P_{high}=0.10). The increased bacterial production was cross-correlated with quadruplicate measurements of biomass from each treatment, and resulting growth rate coefficients (I_{maxH}) for all treatments were within a narrow range (0.359 to 0.550 day^{-1}) and there was no statistically significant difference in growth rates between each nutrient treatment (Fig. 2b) ($P_{low\text{-}medium}$=0.55, $P_{medium\text{-}high}$=0.49, $P_{none\text{-}high}$=0.10). The maximum measured growth rate for a single nutrient treatment, thus equating to the parameter I_{maxH}, was 0.55 day^{-1}. The 95% confidence range for I_{maxH} is 0.50 to 0.60 day^{-1}. This value is, to our knowledge, is the first measured rate of bacterial growth from High-Arctic soils, and falls within the lower end of the plausible range established in Bradley et al. (2015) (0.24 – 4.80 day^{-1}) (Fig. 3a) for soil microbes from a range of laboratory and modelling studies (Frey et al., 2010;

Ingwersen et al., 2008; Knapp et al., 1983; Zelenev et al., 2000; Stapleton et al., 2005; Darrah, 1991; Blagodatsky et al., 1998; Vandewerf and Verstraete, 1987; Foereid and Yearsley, 2004; Toal et al., 2000; Scott et al., 1995). For respiration, significantly higher CO_2 headspace concentration were detected in the live incubations at 25˚C relative to killed controls ($P < 0.05$). Average respiration rate at 5°C was 1.61 µg C g^{-1} day^{-1} and there was a significant increase in soil respiration at 25°C (12.83 µg C g^{-1} day^{-1}) (Fig. 2c) ($P < 0.05$). The Q_{10} value for Midtre Lovénbreen forefield soils was thus calculated as 2.90, and a 95% confidence range was established as 2.65 to 3.16. This was at the upper end of the plausible range previously identified in Bradley et al. (2015) (Fig. 3b). Based on measured values of bacterial production and respiration, BGE (Y_H) was 0.06, with a 95% confidence range of 0.05 to 0.07 (Fig. 3c). Final calculated values for model parameters are summarized in Table S3 (Supplementary Information).

The results from microscopy determination of biomass are presented in Table 2. In the freshly exposed soil (year 0) heterotrophic biomass was low (0.059 µg C g^{-1}), increased substantially to 0.244 µg C g^{-1} in 29 year old soils, and was an order or magnitude higher (2.00 µg C g^{-1}) in 113 year old soils. Autotrophic biomass was considerably higher than heterotrophic biomass and increased by roughly an order of magnitude from year 0 (0.171 µg C g^{-1}) to year 29 (1.07 µg C g^{-1}) and approximately doubled by year 113 (2.58 µg C g^{-1}). TOC in freshly exposed soil was approximately 0.793 mg C g^{-1}).

16S data was categorized into microbial groups (A_{1-3} and H_{1-3}) as defined by the model formulation. Chemolithoautotrophs, such as known iron or sulphur oxidizers (genera *Acidothiobacillus*, *Thiobacillus*, *Gallionella*, *Sulfurimonas*) were assigned into the A_1 group. Phototrophic microorganisms, such as cyanobacteria (*Phormidium*, *Leptolyngbya*) and phototrophic bacteria (*Rhodoferax*, *Erythrobacter*, *Halomicronema*) were allocated into group A_2, while heterocyst forming cyanobacteria from the orders Nostocales and Stigonematales were assigned to group the A_3 (nitrogen-fixing autotrophs). Members of the family Comamonadaceae of the Betaproteobacteria are known glacial dwelling microorganisms (Yde et al., 2010) and were thus included into the group H_1. General soil heterotrophic microorganisms (mainly members of Alphaproteobacteria, Actinobacteria, Bacterioidetes and Acidobacteria) were assigned into group H_2 (general soil heterotrophs). Lastly, group H_3 consisted of heterotrophic nitrogen fixers, mainly *Azospirillum*, *Bradyrhizobium*, *Devosia*, *Clostridium*, *Frankia* and *Rhizobium*. Pathogens, non-soil microorganisms and organisms with unknown physiological traits were assigned into the "Uncategorized" group. Glacial microbes accounted for 43 to 45 % of reads in year 0 and 5, and declined in older soils (year 50 and 113) to 18 to 22%. The glacial community was predominantly chemolithoautotrophic (A_1). Typical soil bacteria (A_2 and H_2) increased from low abundance (30% and 40% in years 0 and 5 respectively) to relatively high abundance (63 to 67% of reads) in years 50 and 113. Nitrogen fixing bacteria were prevalent in recently exposed soils (14% in year 0) but low in relative abundance in soils above 5 years of age (4 to 6% in years 5, 50 and 113). In the freshly exposed soil (year 0), the microbial community was relatively evenly distributed between heterotrophs (43%) and autotrophs (44%). In developed soils, the relative abundance of heterotrophs increased (up to 74% of

reads in years 50 and 113). Important to note is the fact that between 8 and 21% of the reads across all samples could not be classified.

3.2. Model Results

The model predicted an accumulation of autotrophic and heterotrophic biomass over 120 years (Fig. 4a and 4b). Biomass and nutrient concentrations were initially extremely low (total biomass < 0.25 µg C g^{-1}, DIN < 4.0 µg N g^{-1}, DIP < 3.0 µg P g^{-1}), and biological activity in initial soils was also low (Table 3). There was an order of magnitude increase in total microbial biomass in years 10 to 60. Nitrogen-fixing autotrophs (A_3) and heterotrophs (H_3), and soil heterotrophs (H_2) experienced rapid growth during this period. Glacial and soil autotrophs (A_{1-2}) and glacial heterotrophs (H_1) remained low. Bacterial production increased by roughly two orders of magnitude (Table 3). Organic carbon (labile and refractory) increased (Fig. 4c), whilst DIN and DIP concentrations increased by approximately an order of magnitude in the first 60 years (Fig. 4d). During the later stages of soil development (years 60 to 120), biomass increased rapidly due to the rapid growth of soil organisms (A_2 and H_2), which outcompeted nitrogen-fixers. The model showed a rapid exhaustion of labile organic carbon (years 50 to 100), while refractory carbon accumulated slowly. Nutrients (DIN and DIP) accumulated at a relatively constant rate. Microbial activity, including bacterial production, nitrogen fixation and DIN assimilation, was high relative to early stages (Table 3).

A carbon budget of fluxes through the substrate pool is presented in Fig. 5. Daily fluxes are presented in panels (a) for year 5, (b) for year 50 and (c) for year 113, and annual fluxes up to year 120 are presented in (d). In recently exposed soils (5 years), allochthonous inputs were the only noticeable carbon flux, outweighing heterotrophic growth and respiration, and the contribution of substrate from necromass and exudates by over two orders of magnitude (Fig. 5a). Thus, the total change in carbon (black line) closely resembled allochthonous input. In the intermediate stages (Fig. 5b), there was substantial depletion from the substrate pool due to heterotrophic activity. Heterotrophic growth (red line) was low despite high substrate consumption and respiration (dark blue line). In the late stages of soil development, the flux of microbial necromass was a significant contributor to the organic substrate pools (Fig. 5c). Carbon fluxes in mid to late stages of soil development were highly seasonal (Fig. 5b and 5c). Biotic fluxes (e.g. respiration) were up to six times higher during the summer (July to September) compared to the winter (November to April), however a base rate of heterotrophic respiration and turnover of microbial biomass was sustained over winter. Figure 4d shows that the contribution of microbial necromass rose steadily throughout the simulation (light blue line), however was not sufficient to compensate the uptake of carbon substrate, thus leading to overall depletion between years 50 to 110 (black line). The contribution of exudates (purple line) to substrate was minimal at all soil ages.

4. Discussion

4.1. Determination of parameters and model predictions

Figure 6 illustrates the influence of the site-specific, laboratory-derived parameters on microbial biomass predictions. It compares the range of predicted microbial biomass based on laboratory-determined parameters (yellow) to the entire plausible parameter range (red; Bradley et al. (2015)). Predicted biomass with the average laboratory-derived value is indicated by the black line. For I_{max}, predicted biomass with laboratory-derived parameters (yellow shading) was towards the lower end of the plausible range (Fig. 6a) because refined growth rates were significantly lower than the maximum values explored previously. This was mostly due to a significant reduction in autotrophic biomass (A_{1-3}). With high growth rates, there was a sharp early increase in biomass (years 10 to 20) followed by slower growth phase (years 20 to 120). Model results with laboratory-derived growth rates showed that the exponential growth phase occurred later (years 40 to 80) and was more prolonged, but total biomass was considerably lower. There was a substantial reduction in the plausible range in predicted microbial biomass.

There was a substantial reduction in the plausible range in predicted microbial biomass (Fig. 6b) from the measured temperature sensitivity (Q_{10}) (yellow) compared to the previous range (red). Soil microbial communities in Polar regions must contend with extremely harsh environmental conditions such as cold temperatures, frequent freeze-thaw cycles, low water availability, low nutrient availability, high exposure to ultraviolet radiation in the summer, and prolonged periods of darkness in winter. These factors profoundly impact their metabolism and survival strategies and ultimately shape the structure of the microbial community (Cary et al., 2010). High Q_{10} values, as derived here, are typical of cold environments and cold adapted organisms and this has been associated with the survival of biomass under prolonged periods of harsh environmental conditions (Schipper et al., 2014). An investigation into the metabolism of microbial communities in biological soils crusts in recently exposed soils from the Austre Broggerbreen Glacier, approximately 6 km from the Midtre Lovénbreen catchment, also derived a high Q_{10} (3.1) (Yoshitake et al., 2010). The Midtre Lovénbreen catchment, in Svalbard, experiences a relatively extreme Arctic climate. The high Q_{10} ultimately lowers the overall rate of biomass accumulation in ultra-oligotrophic soils and a baseline population is maintained.

The low measured BGE (0.06) suggested that a high proportion (94%) of substrate consumed by heterotrophs is remineralized (degrading organic substrate into DIC (CO_2), DIN and DIP), with very little being incorporated into biomass (6%). Low BGE encouraged the liberation and release of nutrients to the soil and thus the overall growth response of the total microbial biomass was more rapid due to higher soil nutrient concentrations (Fig. 6c). However, due to the low BGE, there was a high rate of substrate degradation, and as such, labile substrate was rapidly depleted when heterotrophic biomass was high (Fig. 4c). Heterotrophic growth requires that a substantial amount of substrate is degraded – thus, although autotrophic production outweighed heterotrophic production at all stages of development (Fig. 4e), the soil was predicted by the model to be a net source of CO_2 to the atmosphere over the first 120 years of exposure (Fig. 4f). Heterotrophic growth and respiration (and thus net ecosystem production and carbon fluxes) are strongly dependant on the availability of soil organic carbon. Poorly quantified rates of allochthonous organic carbon deposition and its quality may lead to generally high

uncertainty in the net ecosystem production due to potentially enhanced heterotrophic growth resulting from higher organic carbon deposition, or lower heterotrophic growth resulting from substrate limitation in low-deposition scenarios. Soil CO_2 efflux is highly sensitive to variable net ecosystem production, thus simulated net ecosystem production estimates must be interpreted cautiously until sufficient field data emerges (e.g. from in situ measurement of soil gas exchange). The calculation of BGE assumes that bacterial respiration is the major contributor to measured CO_2 gas exchange rates from soil microcosms. In reality, all active and living soil organisms are likely to contribute to measured CO_2 fluxes, however due to limitations with experimental protocols, it is extremely difficult to determine the relative contribution of various organisms to total respiration. Microscopy analysis showed limited presence of fungi and protozoa suggesting that the biological community of the soil community is mainly bacterial. Nevertheless, by attributing total measured CO_2 fluxes solely to bacteria, BGE may be under-estimated (due to an overestimation of respiration rates attributed to the bacterial community). Thus, we cannot exclude that our low BGE values might be in part an artefact of this experimental limitation. However, although there are very few measurements of BGE in cold glaciated environments, our estimate of BGE is in good agreement with previous studies, which have suggested values ranging between 0.0035 and 0.033 (Anesio et al., 2010; Hodson et al., 2007). Therefore, we are confident that BGE values measured here fall within a realistic range.

Three assumptions are made in the assignment of measured parameters to SHIMMER as applied to the High-Arctic field site. The first assumption of SHIMMER is that parameter values remain constant throughout the duration of the simulation. Empirical evidence suggests that parameters defined as fixed in SHIMMER (e.g. Q_{10}) may be variable over time, however in SHIMMER, like many numerical modelling formulations, changing environmental (temperature, light) and geochemical (carbon substrate, available nitrogen, available phosphorus) conditions drive subsequent variability in microbial activity via mathematical formulations (e.g. Monod kinetics, see Bradley et al. (2015)) affixed to parameter values. A second assumption is the assignment of measured rates to parameters for all microbial functional groups. Rather than taxonomic based classification, SHIMMER distinguishes and classifies microbial communities based on functional traits. These mathematical formulations assigned to, for example, microbial growth, are different between groups to represent distinct functional traits associated with that group. Whilst actual rates may be different between different organisms, for the level of model complexity and outputs required, a community measurement of those parameters is sufficient, particularly considering that the differences are accounted for in the mathematical formulation of SHIMMER (see Bradley et al. (2015)). Third, maximum microbial growth rate at T_{ref} (25°C, Bradley et al. (2015)) as modelled in SHIMMER is modified by Monod terms that account for nutrient limitation (e.g. Monod terms), as well as a temperature response function (Q_{10}) to estimate actual growth rate at ambient temperature. A major objective of of this study was to improve model performance by constraining previously identified key model parameters (see sensitivity study results in Bradley at al. (2015)) through specifically designed laboratory experiments. We showed this by comparing model simulation results applying measured, site-specific parameter with simulation results using a range of parameter values reported in the literature (Fig. 6).

557

558 **4.2. Microbial biomass dynamics and community structure**

559 Measured microbial biomass in the initial soils of Midtre Lovénbreen (0.23 µg C g^{-1}, 0 years) was very
560 low compared to initial soils in other deglaciated forefields of equivalent ages in lower latitudes, for
561 example in the Alps (4 µg C g^{-1}) (Bernasconi et al., 2011; Tscherko et al., 2003) and Canada (6 µg C g^{-1})
562 (Insam and Haselwandter, 1989). However, our microbial biomass values are more similar to other
563 recently deglaciated soils in Antarctica (Ecology Glacier - 0.88 µg C g^{-1}) (Zdanowski et al., 2013). Low
564 biomass is possibly a result of the harsh, ultra-oligotrophic and nutrient limiting environment of the High
565 Arctic and Antarctica, where low temperature and longer winters limit the summer growth phase,
566 especially compared to an Alpine system (Tscherko et al., 2003; Bernasconi et al., 2011).

567

568 The initial microbial community structure in our samples was predominantly autotrophic (74.5%). In the
569 years following exposure, we observed an increase in autotrophs and heterotrophs with soil age (Table
570 2), presumably due to the establishment and growth of stable soil microbial communities (Schulz et al.,
571 2013; Bradley et al., 2014). Both the observations and modelling results suggested that there was no
572 substantial increase in heterotrophic biomass during the initial and early-intermediate stages of soil
573 development (years 0 to 40), which was then followed by a growth phase whereby biomass increased
574 by roughly an order of magnitude. Overall, the model and the microscopy data were in good agreement
575 accounting for the limitations in both techniques, spatial heterogeneity, and the oscillations in biomass
576 arising from seasonality (Fig. 7). SHIMMMER predicted that low initial microbial populations have the
577 potential to considerably increase in population density during several decades of soil development.
578 This data thus supports the hypothesis that the observed increase in microbial biomass with soil age is
579 a consequence of *in situ* growth and activity. The pattern of microbial abundance observed in the Midtre
580 Lovénbreen forefield broadly resembles that of other glacier forefields worldwide (see Bradley et al.
581 (2014)). For example, data from the Rootmoos Ferner (Austria) (Insam and Haselwandter, 1989),
582 Athabasca (Canada) (Insam and Haselwandter, 1989), Damma (Switzerland) (Bernasconi et al., 2011;
583 Schulz et al., 2013) and Puca (Peru) (Schmidt et al., 2008) glacier forefields find increased microbial
584 biomass and activity over decades to centuries of soil development following exposure.

585

586 The genomic data indicated that glacial microbes (such as members of the family Comamonadaceae.)
587 are dominant in recently exposed soils, in agreement with model results (Fig. 8). The community
588 structure in year 5 was heavily dominated by chemolithoautotrophs (A_1) (including taxa *Thiobacillus*,
589 *Rhodoplanes*, *Acidithiobacillus*, *Nitrospira*, *Sulfurimonas* and others), which reflected findings from
590 previous studies whereby chemolithoautotrophic bacteria contribute to the oxidation of FeS_2 in
591 proglacial moraines in Midtre Lovénbreen (Borin et al., 2010; Mapelli et al., 2011). These processes are
592 also commonly described in other subglacial habitats (Boyd et al., 2014; Hamilton et al., 2013). Based
593 on 16S data, the glacial community declined in relative abundance with soil age. This finding was also
594 reflected in the model in years 50 and 113. As the age of the soil progressed, there was typically greater
595 abundance of microbes representing typical soil bacteria (groups A_2 and H_2 including taxa *Geobacter*,
596 *Micrococcus*, *Actinoplanes*, *Sphingomonas*, *Pedobacter* and *Devosia*, *Frankia*, *Rhizobium*) in the 16S

data and the model, thus the relative abundance of glacial microbes decreased. Relative abundance of microbial communities across the chronosequence is plotted at the phylum and genus level in the Supplementary Information (Fig. S4 and S5). The overall trends show the relative increase in the proportion of Acidobacteria with the soil age. They contain typical soil bacteria and are thus often used as markers of soil formation and soil development. They are usually associated with plant covered older soils with lower pH as they specialize in degradation of plant recalcitrant organic compounds. The younger soils, on the other hand contained relatively higher proportion of sequences of Proteobacteria (particularly Betaproteobacteria), Bacterioidetes and Cyanobacteria, i.e. groups often associated with supra or subglacial habitats.

Microscopic analyses indicated low total biomass in recently exposed soils (up to 1.7 μg C g^{-1} in soil exposed for 50 years) that was comprised predominantly of autotrophic bacteria. Model simulations agreed well with microscopy derived data. Overall, the 16S data, when categorised into functional groups as defined by the model, agreed well with the microscopy and model output in the very early stages of soil development. However, in later stages of soil development (50 years and older), microscopy and modelling suggested a continuation of predominantly autotrophic soil microbial communities whereas 16S sequence data notably indicated a predominantly heterotrophic community. With extremely low biomass, cell counts derived from microscopy, as well as representation of relative abundance by 16S extraction and amplification, can be largely skewed by relatively small changes in the soil microbial community. Furthermore, the comparative difficulty to lyse autotrophic bacteria (such as some groups of cyanobacteria) from an environmental sample compared to heterotrophic bacteria, and thus successfully amplify the 16S gene during the PCR process, may skew 16S sequence data in favour of heterotrophic sequence reads. Incomplete sample depth (shown by rarefaction curves, Fig S6), the differential extractability of DNA from different organisms, and variation in rDNA copy number adds potential for error to the presented 16S data, and thus absolute numbers should be treated with caution. ~~Additionally,~~ SHIMMER is an ambitious model in that it attempts to simulate, predict and constrain multiple functional types of bacteria species in a numerical framework. Numerical models containing multiple species or multiple microbial functional groups are often extremely challenging to constrain (Servedio et al., 2014; Hellweger and Bucci, 2009; Jessup et al., 2004; Larsen et al., 2012), and as such, the majority of microbial soil models often only resolve one or two living biomass pool that represents the bulk activity and function of the entire community (see e.g. Manzoni et al. (2004), Manzoni and Porporato (2007), Blagodatsky and Richter (1998), Ingwersen et al. (2008), Wang et al. (2014) and others). Our rationale for resolving six distinct functional groups was to quantitatively assess, using modelling, the relative importance and role of each functional group at different stages of soil development. Regardless of discrepancies in older soils (over 50 years since exposure), both the 16S and microscopy data indicated that there was a mixed community of autotrophs and heterotrophs in soils of all ages, which was supported by modelling, since no functional groups were extirpated over simulations representing 120 years of soil development. Thus, SHIMMER is able to capture the diversity of the samples over 120 years of soil development, but the detailed community composition requires further investigation.

Nitrogen-fixing bacteria such as *Nostoc, Rivularia, Pseudanabaena* and *Rhodobacter* were prevalent in recently exposed soils but declined in relative abundance with soil age. By fixing N_2 instead of assimilating DIN, the model predicted that nitrogen-fixers were able to grow rapidly in the early stages relative to other organisms (Fig. 4a, 4b). The model prediction supports findings by previous studies demonstrating the importance of nitrogen fixation in Alpine (Duc et al., 2009; Schmidt et al., 2008) and Antarctic (Strauss et al., 2012) glacier forefields and other High-Arctic (Svalbard, Greenland) glacial ecosystems (Telling et al., 2011; Telling et al., 2012). However, there was poor agreement on the relative abundance of nitrogen-fixers between the model and the 16S data in the later stages of soil development (years 50 to 120), particularly between autotrophs and heterotrophs. The model over-predicted the relative abundance of nitrogen-fixing organisms (Fig. 8). The majority of the biomass of the autotrophic nitrogen-fixers was composed of sequences belonging to the cyanobacterium from the genus *Nostoc*. *Nostoc* forms macroscopically visible colonies that grow on the surface of the soils. Its distribution in the Arctic soils is thus extremely patchy and therefore, part of the discrepancy between the 16S data and the model regarding the relative distribution of the A_3 group in the older soils could be due to under-sampling of the *Nostoc* colonies as a consequence of a random sampling approach. Furthermore, allochthonous inputs of nitrogen to the Arctic (e.g. aerial deposition (Geng et al., 2014)) strongly affect the productivity of microbial ecosystems and the requirement of nitrogen fixation for microbes (Bjorkman et al., 2013; Kuhnel et al., 2013; Kuhnel et al., 2011; Hodson et al., 2010; Telling et al., 2012; Galloway et al., 2008). Thus, uncertainty in the allochthonous availability of nitrogen strongly affects nitrogen fixation rates. In attempting to replicate a qualitative understanding of the nitrogen cycle in a quantitative mathematical modelling framework, the predicted importance of nitrogen-fixing organisms may be over-estimated. The poor agreement in the relative abundance of nitrogen-fixers between the model and the 16S data indicates an incomplete understanding of allochthonous versus autochthonous nutrient availability. Allochthonous nutrient availability is a known source of uncertainty (Bradley et al., 2014; Schulz et al., 2013; Schmidt et al., 2008), and addressing this concern is the subject of future work.

16S data is an exciting resource of information that is rarely (or never) used to test numerical process-based biogeochemical models. However, the environment (difficulty to extract DNA), the presentation (percentages of low concentration and thus easy to shift relative abundance), the potentially high proportion of dead or dormant cells (which may be present in sequence data but are not necessarily metabolically active), the differential extractability of DNA from different organisms and variation in rDNA copy number, and uncertainties in model formulation make comparisons challenging. In making this first attempt at comparison of model output to 16S data, we hope to spark discussion and further development of approaches that have similar objectives in order to improve future model performance.

4.3. Net ecosystem metabolism and carbon budget

Allochthonous carbon inputs were the most significant contributor to recently exposed soils (e.g. year 5), since the total change in substrate closely followed this flux (Fig. 5). In older soils (year 113), biotic fluxes were substantially higher, and microbial necromass contributed equally as a source of organic substrate compared to allochthonous deposition. In the older soils, heterotrophic growth and respiration caused substantial consumption and thus depletion of available carbon stocks. This evidence thus supports the hypothesis that carbon fluxes in very recently exposed soils are low and are dominated by abiotic processes (i.e. allochthonous deposition), whereas biotic processes (such as microbial growth, respiration and cell death) play a greater role in developed soils with increased microbial abundance and activity. These findings for the Midtre Lovénbreen glacier in the High-Arctic, are similar to what has been observed based on empirical evidence from Alpine settings (at the Damma Glacier, Switzerland (Smittenberg et al., 2012; Guelland et al., 2013)).

The seasonality of carbon fluxes predicted by the model (Fig. 5b and 5c) related to the high measured Q_{10} values. High seasonal variation in biotic fluxes and rates is typical of cryospheric soil ecosystems (Schostag et al., 2015) including Alpine glacier forefield soils (Lazzaro et al., 2012; Lazzaro et al., 2015). However, microbial activity has been shown to persist during winter under insulating layers of snow and in sub-zero temperatures (Zhang et al., 2014). Modelling also predicted sustained organic substrate degradation, microbial turnover and net heterotrophy during the winter (Fig. 5b and 5c), as documented in other glacier forefield studies from an Alpine setting (Guelland et al., 2013b).

The low measured BGE has three important consequences. Firstly, low BGE suggests that a large pool of substrate is required to support heterotrophic growth. Low-efficiency heterotrophic growth lead to the rapid depletion of substrate; therefore high allochthonous inputs were required to maintain a sizeable pool. In older soils (years 80 to 120), increased inputs from microbial necromass (blue line, Fig. 5d) sustained substrate supply to heterotrophs. The sources of allochthonous carbon substrate to the glacier forefield include meltwater inputs derived from the supraglacial and subglacial ecosystems (Stibal et al., 2008; Hodson et al., 2005; Mindl et al., 2007), snow algae (which are known to be prolific primary colonizers and producers in High Arctic snow packs (Lutz et al., 2015; Lutz et al., 2014), atmospheric deposition (Kuhnel et al., 2013) and ornithogenic deposition (e.g. faecal matter of birds and animals) (Jakubas et al., 2008; Ziolek and Melke, 2014; Luoto et al., 2015; Michelutti et al., 2009; Michelutti et al., 2011; Moe et al., 2009). Microbial dynamics are moderately sensitive to external allochthonous inputs of substrate (Bradley et al., 2015), and addressing the uncertainty associated with this flux is an important question to address in future research.

Secondly, low BGE causes a net efflux of CO_2 over the first 120 years of soil development despite high autotrophic production (Fig. 4e and 4f). Recent literature has explored the carbon dynamics of glacier forefield ecosystems, finding highly variable soil respiration rates (Bekku et al., 2004; Schulz et al., 2013; Guelland et al., 2013a). Future studies should focus on quantifying carbon and nutrient transformations and the potential for forefield systems to impact global biogeochemical cycles in response to future climate change (Smittenberg et al., 2012) and in the context of large-scale ice retreat.

Thirdly, high rates of substrate degradation encouraged by low BGE were responsible for rapid nutrient release. Modelling suggested that microbial growth was strongly inhibited by low nutrient availability in initial soils (4 µg N g^{-1}, 2 to 10 µg P g^{-1}) (Fig. 4d). This is consistent with findings from the Hailuogou Glacier (Gongga Shan, China) and Damma Glacier (Switzerland) (Prietzel et al., 2013). Low BGE is predicted by the model to have a very important role in encouraging the release of nutrients from organic material more rapidly, thereby increasing total bacterial production in the intermediate stages of soil development. Increased nutrient availability with increased heterotrophic biomass is consistent with recent observations from glacier forefields (Bekku et al., 2004; Schulz et al., 2013; Schmidt et al., 2008).

5. Conclusions

We used laboratory-based mesocosm experiments to measure three key model parameters: maximum microbial growth rate (I_{max}) (by incorporation of ^{3}H-leucine), BGE (Y) (by measuring respiration rates) and the temperature response (Q_{10}) (by measuring rates at different ambient temperatures). Laboratory-derived parameters were comparable with previous estimations. We refined model predictions constraining site-specific parameters by lab experiments, thus decreasing parameter uncertainty and narrowing the range of model output over nominal environmental conditions. A comparison of model simulations using laboratory-derived parameter values and previously defined parameter values showed that the coupling of high Q_{10} values and low BGE were important factors in controlling biomass accumulation due to promoting survival of biomass during periods of low temperature, and the enhanced recycling of nutrients through organic matter degradation, respectively. Our results demonstrated that *in situ* microbial growth lead to the overall accumulation of microbial biomass in the Midtre Lovénbreen forefield during the first century of soil development following exposure. Furthermore, carbon fluxes increased in older soils due to elevated biotic (microbial) activity. Microbial dynamics at the initial stages of soil development in glacial forefields do not contribute to significant accumulation of organic carbon due to the very low growth efficiency of the microbial community, resulting in a net efflux of CO_2 from those habitats. However, the low bacterial growth efficiency in glacial forefields is also responsible for high rates of nutrient remineralization that most probably has an important role on the establishment of plants at older ages. The relative importance of allochthonous versus autochthonous substrate and nutrients is the focus of future research.

This exercise shows how an integrated model-data approach can improve understanding and predictions of microbial dynamics in forefield soils and disentangle complex process interactions to ascertain the relative importance of each process independently. This would, for annual budgets, be extremely challenging with a purely empirical approach. Nevertheless, more clarity and data are needed in tracing the dynamics and interactions of these carbon pools to improve confidence and validate model simulations. Proglacial zones are expanding due to accelerated ice retreat. Thus, glacier forefields are becoming an increasingly important novel habitat for microorganisms in glaciated regions experiencing rapid changes in climate. This combined approach explored detailed microbial and biogeochemical

dynamics of soil development with the view to obtaining a more holistic picture of soil development in a warmer and increasingly ice-free future world.

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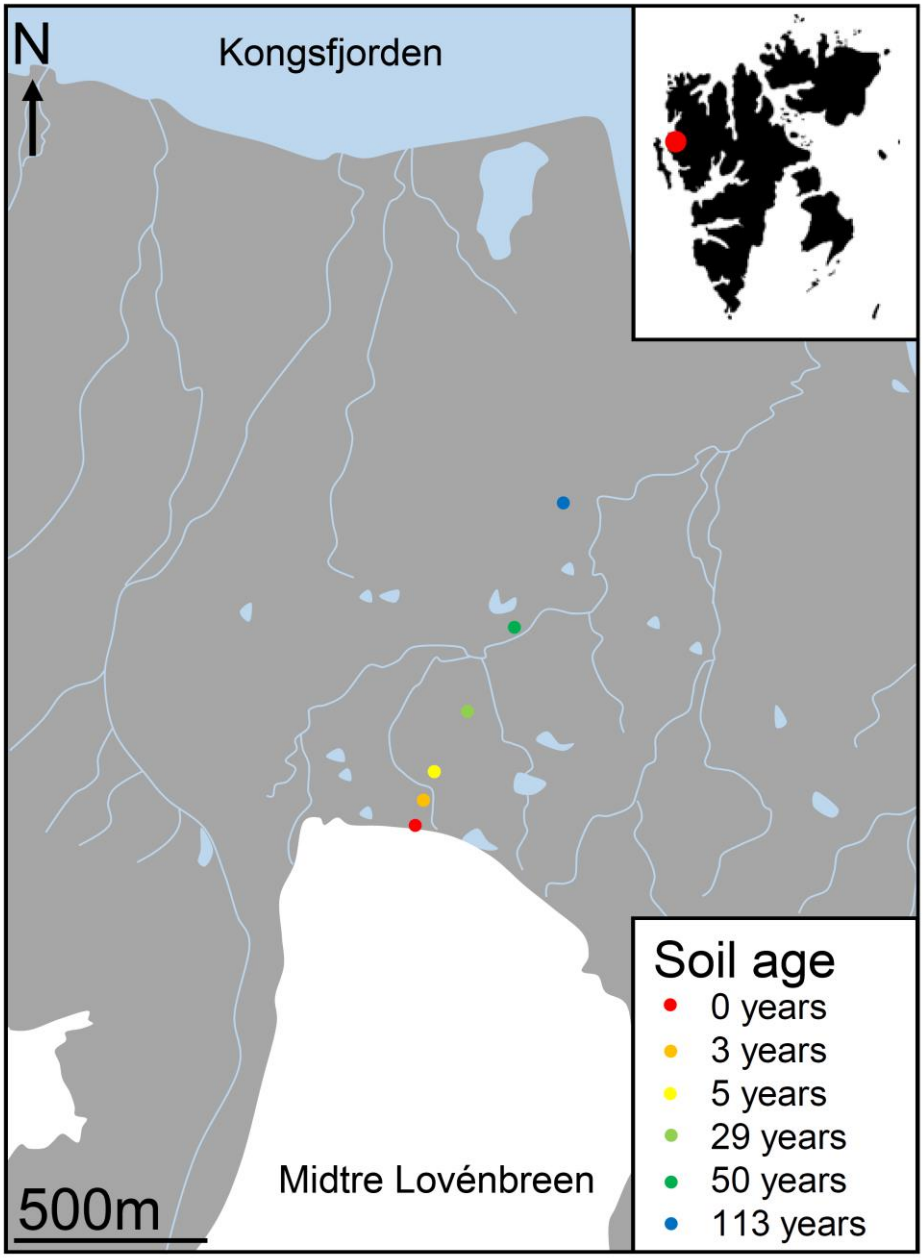

Figure 1. Midtre Lovénbreen glacier and forefield in Svalbard, the location of sampling sites and approximate age of soil.

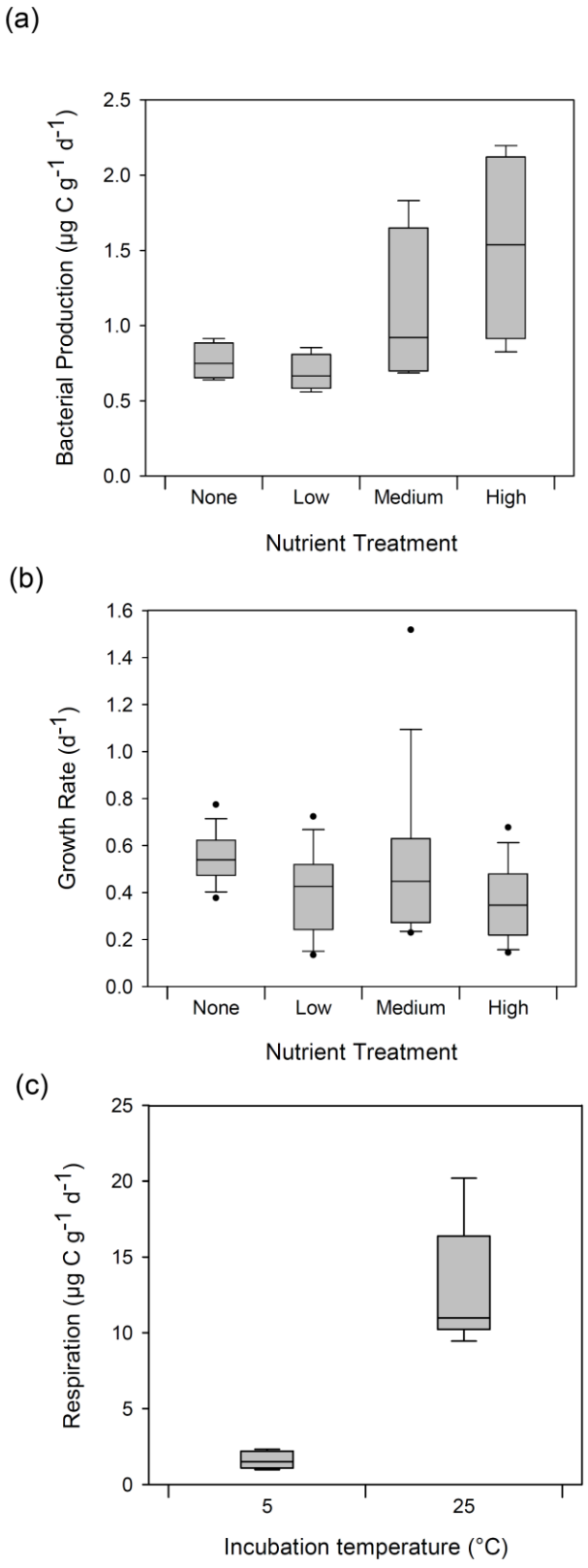

Figure 2. Measurements of (a) bacterial carbon production and (b) growth rate, derived from ^{3}H-leucine assays at different nutrient conditions, and (c) bacterial respiration at 5˚C and 25˚C.

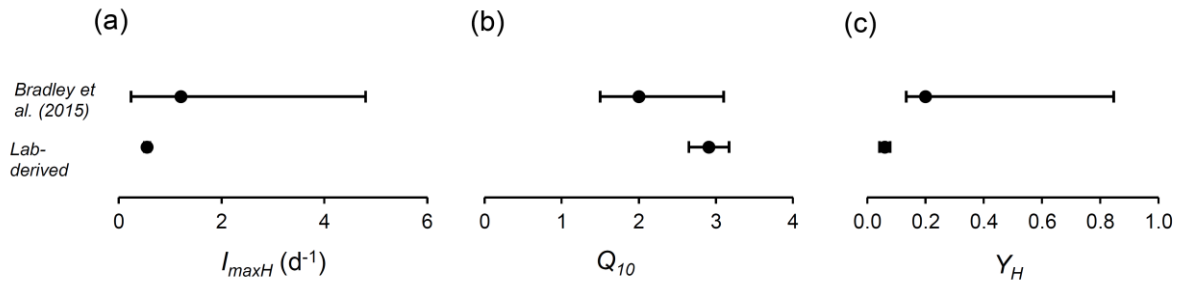

Figure 3. A comparison of previously established ranges for parameters (Bradley et al., 2015) with laboratory-derived values for (a) maximum growth rate (I_{max}), (b) temperature response (Q_{10}), (c) BGE (Y).

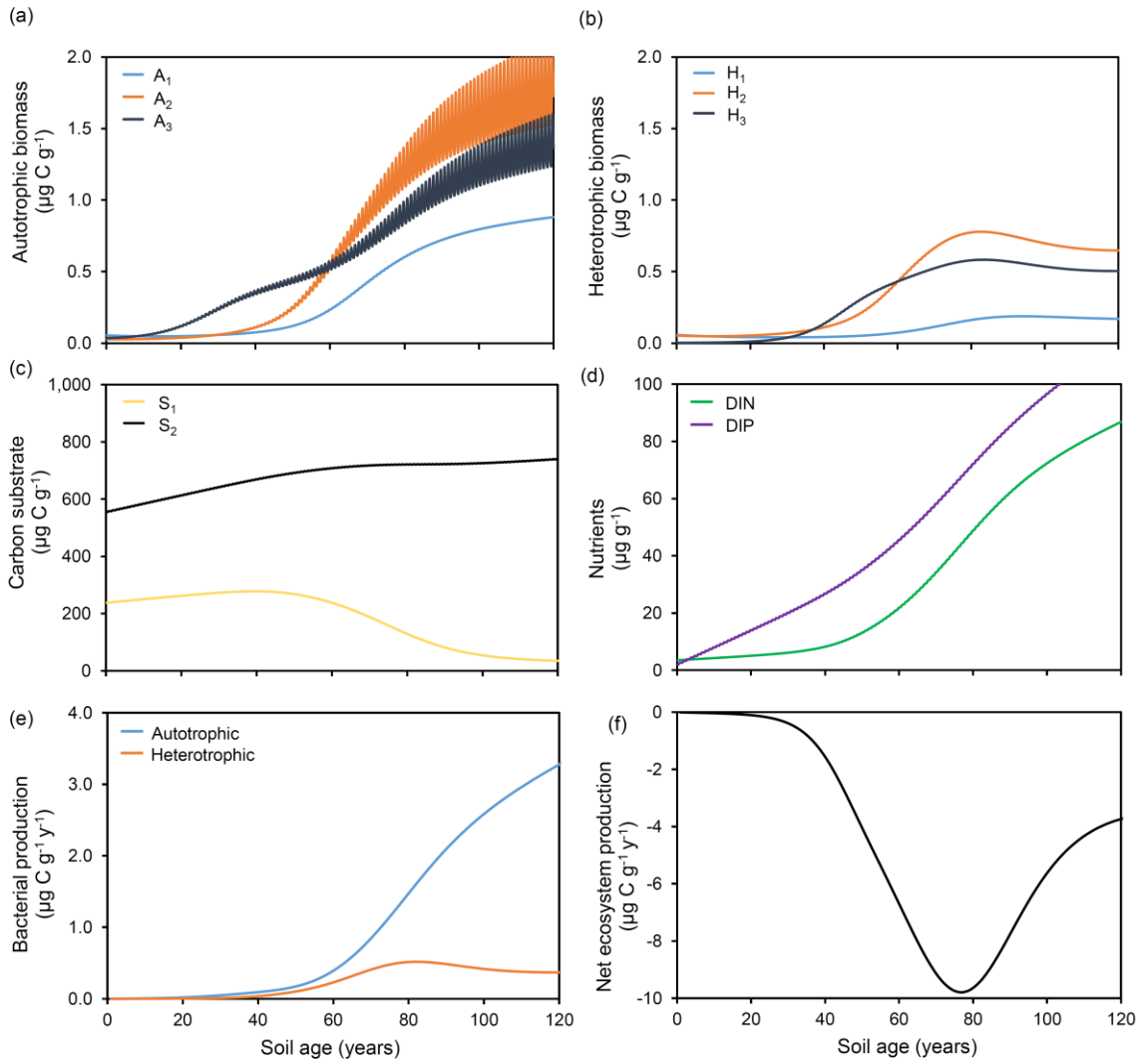

Figure 4. Modelled (a) autotrophic biomass, (b) heterotrophic biomass, (c) carbon substrate, (d) nutrients, (e) bacterial production and (f) net ecosystem production, with laboratory-derived parameter values.

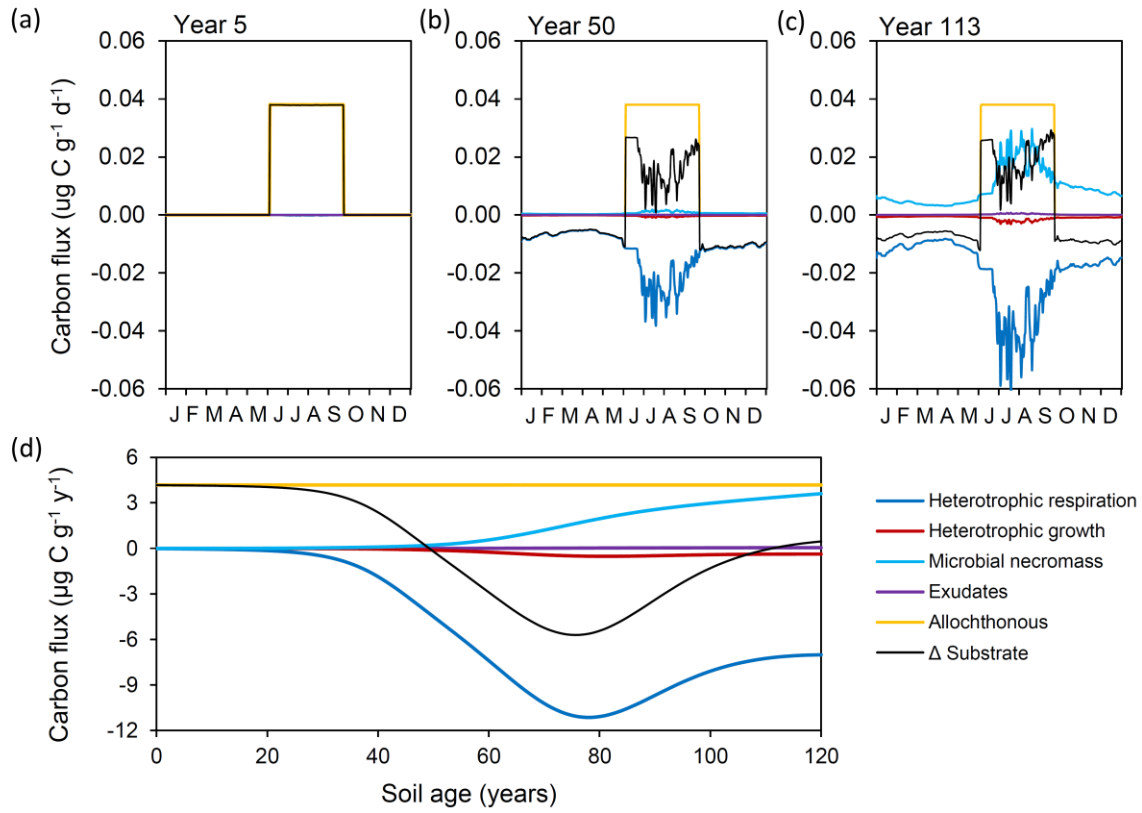

Figure 5. Illustration of daily carbon fluxes for (a) 5, (b) 50 and (c) 113 year old soil, and (d) annual carbon flux over 120 years. Microbial necromass (light blue), exudates (purple) and allochthonous sources (yellow) contribute to the substrate pool (black), and heterotrophic growth (red) and respiration (dark blue) deplete it.

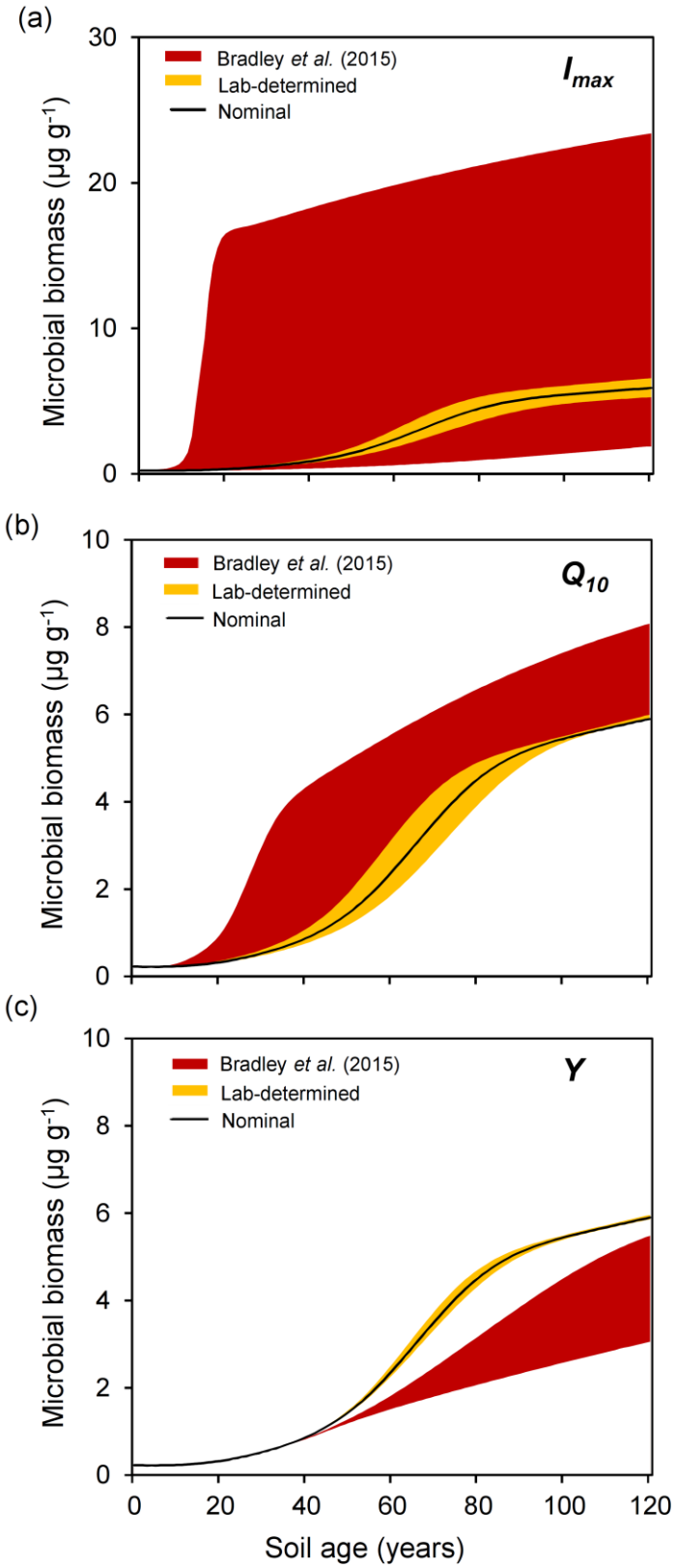

Figure 6. A comparison of predicted microbial biomass with laboratory-derived parameter values (yellow) and previously established parameter values (Bradley et al., 2015) (red) for variation in the following parameters: (a) maximum growth rate (I_{max}), (b) temperature response (Q_{10}), (c) BGE (Y).

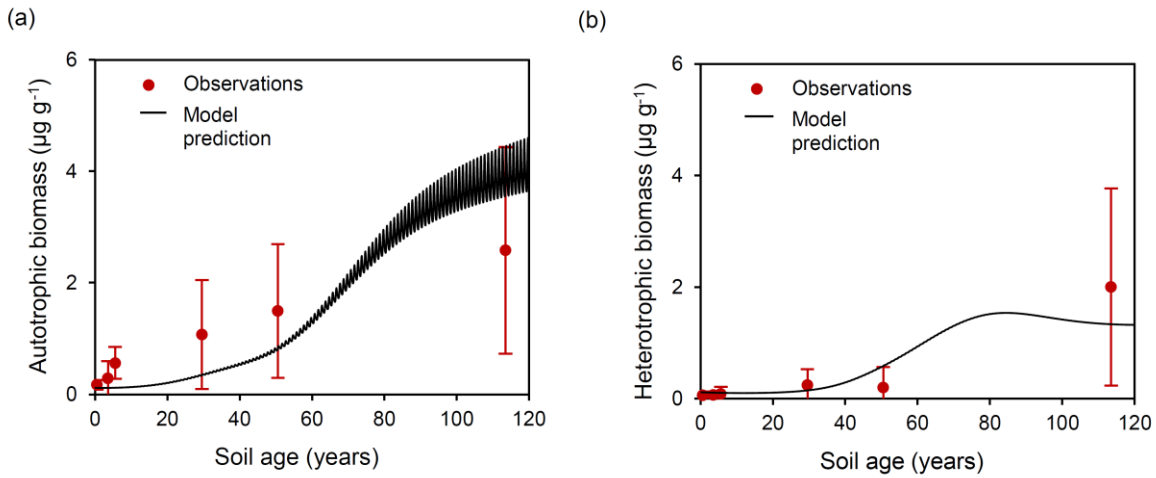

Figure 7. Model predictions of (a) autotrophic and (b) heterotrophic biomass (black line), compared to observational data (red) derived from microscopy.

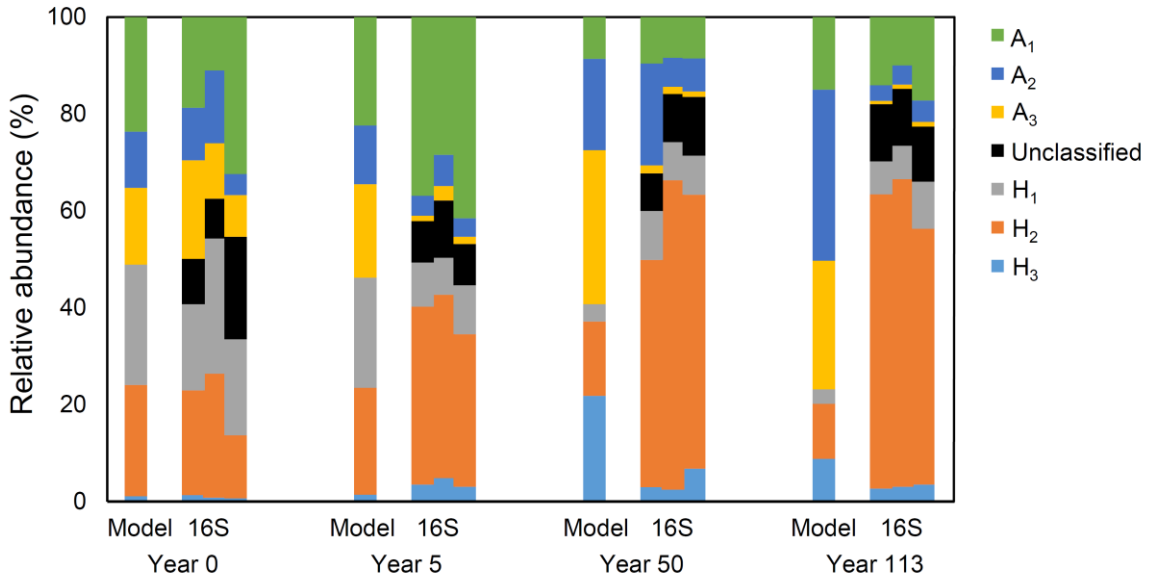

Figure 8. A comparison of microbial diversity from model output and genomic analyses at 0 year old, 5 year old, 50 year old and 113 year old soil.

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1152 Table 1. State variables and initial values.

State Variable	Units	Description	Initial value (year 0) ($\mu g\ g^{-1}$)
A_1	$\mu g\ C\ g^{-1}$	Glacial chemolithoautotrophs	*0.0547*
A_2	$\mu g\ C\ g^{-1}$	Soil autotrophs	*0.0266*
A_3	$\mu g\ C\ g^{-1}$	Nitrogen fixing soil autotrophs	*0.0355*
H_1	$\mu g\ C\ g^{-1}$	Glacial heterotrophs	*0.0576*
H_2	$\mu g\ C\ g^{-1}$	Soil heterotrophs	*0.0530*
H_3	$\mu g\ C\ g^{-1}$	Nitrogen fixing soil heterotrophs	*0.0025*
S_1	$\mu g\ C\ g^{-1}$	Labile organic carbon	*291.895*
S_2	$\mu g\ C\ g^{-1}$	Refractory organic carbon	*681.089*
DIN	$\mu g\ N\ g^{-1}$	Dissolved inorganic nitrogen (DIN)	*3.530*
DIP	$\mu g\ P\ g^{-1}$	Dissolved inorganic phosphorus (DIP)	*2.078*
ON_1	$\mu g\ N\ g^{-1}$	Labile organic nitrogen	*41.157*
ON_2	$\mu g\ N\ g^{-1}$	Refractory organic nitrogen	*96.034*
OP_1	$\mu g\ P\ g^{-1}$	Labile organic phosphorus	*24.227*
OP_2	$\mu g\ P\ g^{-1}$	Refractory organic phosphorus	*56.530*

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1157 Table 2. Microbial biomass in the forefield of Midtre Lovénbreen (brackets show 1 standard deviation)
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Soil Age (years)	Autotrophic biomass (µg C g^{-1})		Heterotrophic biomass (µg C g^{-1})		Total Organic Carbon (µg C g^{-1})	
0	0.171	(0.042)	0.059	*(0.034)*	792.984	(127.206)
3	0.287	(0.155)	0.064	*(0.029)*		
5	0.561	(0.143)	0.083	*(0.065)*		
29	1.072	(0.487)	0.244	*(0.142)*		
50	1.497	(0.601)	0.197	*(0.184)*		
113	2.581	(0.927)	2.000	*(0.885)*		

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Table 3. Model output.

Soil Age (years)	Autotrophic biomass (µg C g^{-1})	Heterotrophic biomass (µg C g^{-1})	Autotrophic production (µg C g^{-1} y^{-1})	Heterotrophic production (µg C g^{-1} y^{-1})	Net ecosystem production (µg C g^{-1} y^{-1})	DIN assimilation (µg N g^{-1} y^{-1})	N_2 fixation (µg N g^{-1} y^{-1})
0	0.117	0.111	0.002	0.001	- 0.011	2.0×10^{-4}	2.0×10^{-4}
3	0.117	0.105	0.003	0.001	- 0.020	3.0×10^{-4}	3.0×10^{-4}
5	0.119	0.102	0.004	0.001	- 0.025	4.0×10^{-4}	4.0×10^{-4}
29	0.359	0.147	0.050	0.012	- 0.391	0.002	0.006
50	0.860	0.591	0.187	0.113	- 4.311	0.022	0.021
113	4.414	1.331	3.093	0.376	- 4.031	0.458	0.031